How Hypergame Theory Is Revolutionizing High-Stakes AI Training

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From random youtube video: https://www.youtube.com/watch?v=bWMjJUr6oW4&t=1s screen shot. Shows armoured ambulances, but with gun ports. Copyright: Canadian company Roshel. Fictive scenario of unknown ambulances with wounded people, requesting access to a guarded base/facility.

How Hypergame Theory Is Revolutionizing High-Stakes AI Training

The hidden crisis: Adversaries with different mental models. The solution: Hypergame-aware AI. When adversaries operate under different rules than you assume and exploit your assumptions, your AI needs to see what is going on

EXECUTIVE SUMMARY

A breakthrough paper from Royal Holloway University introduces the first automated framework for analyzing strategic misalignment in multi-agent systems. Traditional game theory assumes all players

share the same understanding of strategic situations—but real adversaries in cybersecurity, military operations, and business competition operate with fundamentally different mental models. This article explains how Trencsenyi's hypergame rationalization framework solves this problem through formal logic and answer-set programming, enabling post-incident analysis and realistic AI training.

We demonstrate how this theoretical advance integrates with 20+ years of defense sector deployment (General Dynamics), practical operational frameworks (OODA Loop, GFI metrics), cognitive threat modeling (CDCL Framework), formal algorithmic foundations (LegoNE), belief state modeling, and current military AI systems (VR-Forces combat pilots).

For practitioners: Hypergame-aware AI finally provides tools to train against adversaries who don't just challenge tactics—they challenge your fundamental understanding of the game itself. Key applications span cybersecurity threat intelligence, military training, AI safety research, executive decision-making, and autonomous systems.

Reading time: 25–30 minutes for full article | This executive summary: 2 minutes

⚠️ CRITICAL TERMINOLOGY NOTE

Understanding "Misalignment" in This Context

The research paper discussed in this article uses the term **"misalignment"** prominently, even in its title: *"Solving Agent Misalignment In Strategic Play."* However, this refers to something fundamentally different from what AI researchers typically mean by "misalignment" in 2025.

Traditional AI Misalignment (what you're probably thinking of): When **friendly AI systems fail to align with their owner's/user's goals, interests, and mission.** For example, if you ask an AI to polish and check an article you wrote, you don't want it to say "I updated

and corrected the article, it is ready for publish" while *secretly* it hallucinated and fabricated fake research papers under your name and inserted them into your previously legitimate reference list. That would harm your interests and reputation as the author.

Hypergame Misalignment (what this article discusses): **Different strategic mental models between competing agents or adversaries** in game-theoretic scenarios. This refers to situations where each player in a strategic interaction (military conflict, cybersecurity engagement, business competition) operates with a **different subjective understanding of the "game" being played**—different beliefs about available moves, opponent objectives, payoffs, or even who the players are.

Hypergame theory is an expansion of classical Game Theory that explicitly models these divergent mental models. **Critically, hypergames encompass far more than just deception tactics**—they model the full spectrum of differing perceptions of reality:

> **Active Deception:** Honeypot operations (victim thinks the game is 'love'), Trojan Horse gifts (contains offensive payload), hunting calls (false mating signals to lure prey), false flag operations (attacks disguised as coming from a different adversary)
>
> **Information Asymmetry**: Simply being better informed about what is really happening, what the actual game is, who is playing which role, and why they're playing it
>
> **Constructed Realities:** Baudrillard's *Simulacra and Simulation* (famously shown as a book in the movie *The Matrix*, 1999) describes how people perceive false copies of reality—not necessarily through technological simulation (as in Bostrom's work or the film's literal interpretation), but through media, cultural narratives, and selective presentation of facts. News media, for example, present curated selections from particular angles—never the totality of true reality, even when reporting truthfully

This type of misalignment is not a bug; it's the reality of all strategic interactions that we must train for. Understanding and leveraging

hypergame misalignment is essential for realistic high-stakes training in cybersecurity, military operations, intelligence work, and strategic leadership.

The Hidden Crisis in Strategic AI

Imagine training military officers for strategic negotiations, only to discover that your AI opponent thinks it's playing an entirely different game. Or deploying autonomous security systems that misinterpret threats because they fundamentally misunderstand the adversary's objectives. This isn't science fiction—it's the reality of agent misalignment in multi-agent systems, and it's one of the most pressing challenges in AI safety today.

Hypergame Rationalisability: Solving Agent Misalignment In Strategic Play

Differences in perception, information asymmetries, and bounded rationality lead game...

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arXiv > cs > arXiv:2512.11942

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Computer Science > Artificial Intelligence

[Submitted on 12 Dec 2025]

Hypergame Rationalisability: Solving Agent Misalignment In Strategic Play

Vince Trencsenyi

Differences in perception, information asymmetries, and bounded rationality lead game-theoretic players to derive a private, subjective view of the game that may diverge from the underlying ground-truth scenario and may be misaligned with other players' interpretations. While typical game-theoretic assumptions often overlook such heterogeneity, hypergame theory provides the mathematical framework to reason about mismatched mental models. Although hypergames have recently gained traction in dynamic applications concerning uncertainty, their practical adoption in multi-agent system research has been hindered by the lack of a unifying, formal, and practical representation language, as well as scalable algorithms for managing complex hypergame structures and equilibria. Our work addresses this gap by introducing a declarative, logic-based domain-specific language for encoding hypergame structures and hypergame solution concepts. Leveraging answer-set programming, we develop an automated pipeline for instantiating hypergame structures and running our novel hypergame rationalisation procedure, a mechanism for finding belief structures that justify seemingly irrational outcomes. The proposed language establishes a unifying formalism for hypergames and serves as a foundation for developing nuanced, belief-based heterogeneous reasoners, offering a verifiable context with logical guarantees. Together, these contributions establish the connection between hypergame theory, multi-agent systems, and strategic AI.

https://arxiv.org/abs/2512.11942 copyright: arXiv.org and Vince Trencsenyi

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Hypergame Rationalisability: Solving Agent Misalignment In Strategic Play

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Abstract. Differences in perception, information asymmetries, and bounded rationality lead game-theoretic players to derive a private, subjective view of the game that may diverge from the underlying "ground-truth" scenario and may be misaligned with other players' interpretations. While typical game-theoretic assumptions often overlook such

screenshot, copyright: Vince Trencsenyi (the Abstract text continues, this is just a small screenshot of the actual paper)

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A groundbreaking new paper from Royal Holloway University of London offers the first practical solution to this problem. **"Hypergame Rationalisability: Solving Agent Misalignment In Strategic Play"** by Vince Trencsenyi (published last week, submitted to arXiv.org on 12 Dec 2025) introduces a formal, logic-based framework for modeling and analyzing situations where different agents—human or artificial—operate with fundamentally different mental models of strategic situations.

What Makes This Research Revolutionary?

The Core Innovation

Traditional game theory assumes all players share the same understanding of the game they're playing. They know the rules, the available moves, the payoffs, and each other's objectives. But real-world strategic interactions—especially in cybersecurity, military operations, and intelligence—rarely work this way.

Hypergame theory addresses this gap by acknowledging that each player develops a subjective view of the strategic situation based on their:

Information asymmetries

Cognitive limitations

Cultural or institutional biases

Deliberate deception strategies

Bounded rationality

Constructed or filtered perceptions of reality

What Trencsenyi's paper achieves for the first time is:

A Declarative Language: A domain-specific language (DSL) grounded in logic that can formally encode hypergame structures—different players' subjective interpretations of the same strategic situation.

Automated Rationalization: An answer-set programming pipeline that can reverse-engineer belief structures. Given an observed outcome that seems "irrational," the system can automatically identify what beliefs would make that outcome rational from each player's perspective.

Scalable Computation: Unlike previous hypergame approaches that required manual analysis, this framework provides computational tractability for complex multi-agent scenarios.

Verifiable Logic: The system offers logical guarantees—you can verify that an AI agent's reasoning is consistent with its stated beliefs, rather than operating as an unexplainable black box.

The Technical Breakthrough

The paper introduces two key concepts:

Strong Hypergame Nash Equilibrium (s-HNE): A strategy profile where each player's choice is optimal given their subjective view of the game—and that view must be mutually consistent.

Weak Hypergame Nash Equilibrium (w-HNE): A more flexible criterion where each player's choice merely needs to belong to some

Nash equilibrium in their subjective game, allowing for deeper misalignments.

The "umpire" mechanism—a pseudo-player that rationalizes outcomes—can systematically search for belief structures that justify seemingly irrational behaviors. This is revolutionary for post-incident analysis in cybersecurity and military operations.

Real-World Validation: From Historical Analysis to Modern Deployment

The Fall of France: A Classic Hypergame

The paper demonstrates its framework by reconstructing one of history's most catastrophic strategic failures: the 1940 Fall of France. The French High Command reinforced the North while the German Wehrmacht attacked through the supposedly impassable Ardennes forest—a move the French literally didn't believe was possible.

Traditional analysis blamed French incompetence. But hypergame analysis reveals something more nuanced: **the French strategy was actually rational given their subjective game** (which excluded the Ardennes option). The Germans, however, had intelligence suggesting French reinforcement would go North, creating a reduced subgame where attacking through Ardennes became the optimal surprise move.

The automated rationalization framework successfully recovers this belief structure, demonstrating that what appears irrational from an omniscient observer's perspective makes perfect sense when you model each side's actual mental model.

From Theory to Practice: Two Decades of Hypergame Deployment

While Trencsenyi's paper provides the first automated computational framework for hypergame rationalization, hypergame theory itself has been quietly deployed in defense and

intelligence applications for over two decades. Dr. Russell Vane's pioneering work at General Dynamics demonstrated that hypergames aren't just theoretical constructs—they're practical tools for real-world adversarial reasoning.

Vane identified eight critical factors that cause strategic misalignment in competitive situations:

Differences in knowledge and expertise—What one player understands, another may not

Divergent situation assessments—The same battlefield looks different to different commanders

Asymmetric information access—Intelligence gaps create subjective games

Projection capability differences—Understanding 'what beats what' varies by player

Operational information asymmetries—Both at commitment and during execution

Plan robustness variations—Some strategies are brittle; others resilient

Time constraints—Rushed decisions create simplified mental models

Creativity differentials—Deception, feints, and hidden reserves

These factors directly validate why Trencsenyi's automated rationalization framework is essential—the complexity of real-world strategic misalignment requires computational tools to manage.

Defense Industry Implementation: General Dynamics has implemented hypergame-aware Decision-Theoretic/Game-Theoretic (DTGT) agents in operational systems, demonstrating that the technology has matured beyond academic research into deployable solutions. This 20-year track record (Vane, 2006) provides confidence that the computational advances in Trencsenyi's work will find immediate practical application.

The Broader Research Ecosystem: From Formal Languages to Operational Deployment

While Trencsenyi's framework provides automated rationalization of strategic misalignment, my broader research program demonstrates how hypergame theory integrates across five critical dimensions:

1. Formal Algorithmic Foundation: The LegoNE Connection

Recent work on **LegoNE** (Li et al., arXiv:2508.11874) demonstrates how game-theoretic algorithms can be formally specified using domain-specific languages. LegoNE is a Python-like language for defining approximate Nash Equilibrium algorithms through:

> **Operation definitions**: Building blocks like best_response, eqmix, and optimal_mix
>
> **Constraint specifications**: First-order arithmetic formulas governing strategic behavior
>
> **Compiler-based approach**: Automatic translation to constraint programming (e.g., Mathematica)

Connection to hypergames: Trencsenyi's DSL for hypergame structures shares conceptual DNA with LegoNE—both use formal languages to make game-theoretic reasoning machine-processable. Where LegoNE compiles to constraint solvers, Trencsenyi compiles to answer-set programming. Together, they represent the **formal language infrastructure** needed for computational game theory.

Practical implication: Organizations designing hypergame-aware training systems can use LegoNE-style specifications to formally define NPC behavior, ensuring reproducibility and formal verification. My work on **multi-agent collaboration protocols** extends this principle—specialized AI agents (Strategic AI Agent, Generative AI Agent) operate as modular "operations" that combine to produce emergent strategic intelligence.

2. Belief State Modeling: The Cognitive Infrastructure

Recent work by Morenville et al. (**"Modeling Uncertainty: Constraint-Based Belief States in Imperfect-Information Games,"** arXiv, July 2025) provides the computational cognitive infrastructure that enables hypergame-aware AI to function in practice. While Trencsenyi's framework rationalizes beliefs from observed outcomes, Morenville's work shows how AI agents maintain and update belief states in real-time during strategic interactions.

Core Technical Contributions:

Constraint-Based Representation: Uses Constraint Satisfaction Problems (CSPs) to logically deduce the set of feasible opponent states. This is pure logical deduction—starting with premises (all possible states) and applying rules of inference to eliminate contradictions, yielding only states that remain logically consistent with observed information.

Probabilistic Representation: Augments constraint-based models with Belief Propagation (BP) to estimate marginal probability distributions over hidden elements. This moves beyond "possible/impossible" to quantitative likelihood measures—not just "which states are feasible" but "which feasible states are most probable."

Surprising Finding: Experiments reveal that constraint-based belief states often achieve performance comparable to computationally intensive probabilistic inference. For many hypergame scenarios, knowing what's definitely possible (and what's definitely not) provides sufficient information for effective decision-making without quantifying exact probabilities.

Integration with Hypergame Theory:

My synthesis work (**"Advanced AI in Imperfect-Information Environments,"** July 2025) demonstrates how these belief modeling techniques enable hypergame-aware systems to:

> **Track opponent mental models:** Maintain real-time estimates of what adversaries believe about the game state, their

objectives, and available moves

Detect strategic deception: Identify when opponent actions are inconsistent with their purported beliefs, revealing hidden agendas

Generate coherent deceptions: Maintain internally consistent false belief states to project to adversaries while tracking the true game state

This provides the **computational backbone** connecting Trencsenyi's rationalization framework (reverse-engineering beliefs from past outcomes) with real-time strategic reasoning (updating beliefs as interactions unfold). Together, they enable AI systems to operate in the "belief space" where hypergames occur—not just modeling what happened, but predicting and influencing what opponents will do next.

Practical Implication: Organizations implementing hypergame-aware training systems need both components: Trencsenyi's logic-based DSL for encoding hypergame structures and Morenville's belief state algorithms for real-time opponent modeling. The CSP approach's efficiency (comparable performance to probabilistic methods with lower computational cost) makes deployment feasible even in resource-constrained environments.

3. Operational Framework: The Algorithmic Nexus

Trencsenyi's rationalization engine needs an execution framework. My **"Algorithmic Nexus of Strategic Action"** provides this through:

ε-Generative Intelligence Loop (ε-GIL): Extends Boyd's OODA Loop with explicit Evaluate/Adapt phases, where ε measures deviation from desired outcomes.

Generative Fidelity Index (GFI): Quantifies strategic effectiveness as $F=1-\varepsilon$, with ε decomposed into $E_{outcome}$ (goal achievement error), $E_{resource}$ (efficiency error), and $E_{stability}$ (systemic disruption).

CIA Triad Integration: Maps Confidentiality, Integrity, and

Availability as both strategic enablers and attack surfaces across the decision cycle.

Applications:

Internal optimization: Track NPC training effectiveness via trainee ε reduction over time

Competitive intelligence: Infer adversary strategic maturity from observed GFI

Substrate-agnostic: Validated across high-frequency trading, biological systems (lion hunting), and technology markets

4. Threat Modeling: The CDCL Framework

The **Cognitive Deception & Control Landscape (CDCL) Framework** reveals that hypergame strategies represent real threat vectors, not just training abstractions:

Hypergame-Informed Threat Actors (HITA): Advanced AI systems that actively manipulate knowledge graphs to construct deceptive realities for competing agents. Research validates these risks:

Zhao et al. (2024): Untargeted attacks degrade AI understanding globally without knowing specific targets

Zhang et al. (2019): Data poisoning via indirect attacks (multi-hop perturbations) evades detection

Raman et al. (2021): AIs maintain performance while drastically altering semantic structure—deception without observable degradation

Implication for training: Cybersecurity simulations must model adversaries who don't just hide information—they **actively corrupt the defender's perceived reality**. This moves beyond red team exercises to **cognitive warfare simulation**.

5. Military Validation: Operational Deployment

The theory-to-practice transition is underway. **Selmonaj et al. (arXiv:2509.26002, 2025)** demonstrate **AI combat pilots trained via Multi-Agent Reinforcement Learning** deployed in VR-Forces—a standard defense simulation platform.

Key features:

- **Adaptive adversaries**: Not scripted bots, but agents exhibiting "distinct combat behaviors"
- **Real-time human-AI interaction**: Pilots engage intelligent agents in 3D air combat
- **Emergent tactics**: MARL enables discovery of novel strategies

My **EAISE (Embodied AI Simulation Environment)** framework provides the architectural blueprint:

- Multi-agent RL infrastructure
- Self-evolving tactical behaviors
- Mirror testing for self-awareness validation
- **Self-Reflecting Tactician** concept: Integration of strategic deception, transactional precision, and self-aware learning

Key insight: Defense applications aren't waiting for hypergame theory to mature—they're deploying it now. Cybersecurity, intelligence, and executive training must catch up.

Integration: The Complete Stack

These five layers create a comprehensive framework:

Trencsenyi: Automated Rationalization ← Logic-based analysis
LegoNE: Formal Algorithm Specification ← Compilation infrastructure
Belief States: Cognitive Modeling ← Real-time opponent tracking
Algorithmic Nexus: Execution Framework ← Operational metrics (GFI)
CDCL: Threat Modeling ← Adversarial dynamics
EAISE/VR-Forces: Deployment ← Real-world validation

Practical workflow:

Design: Use LegoNE-style formal specifications for NPC behavior

Implement: Deploy Morenville's belief state tracking for real-time modeling

Execute: Run via Algorithmic Nexus framework with GFI tracking

Secure: Model HITA-class threats via CDCL analysis

Train: Implement in EAISE/VR-Forces environments

Analyze: Use Trencsenyi's rationalization to explain observed outcomes

Practical Applications: Where This Matters Now

1. Cybersecurity & Threat Intelligence

Current Challenge: Security teams analyze attacker behavior assuming shared strategic frameworks. But sophisticated adversaries deliberately manipulate these frameworks.

Hypergame Solution:

Model attacker belief structures that rationalize observed behavior

Identify where attackers might misperceive your defenses

Design deception strategies that exploit attacker misalignment

Automate post-incident analysis to uncover strategic assumptions

Example: A ransomware gang's behavior only makes sense if they believe your backup systems are compromised. The rationalization framework can identify this belief, allowing you to either exploit it or

correct your security posture.

2. Military & Strategic Training

Current Challenge: Training scenarios often use opponents that are either too predictable (scripted) or too random (stochastic). Neither prepares personnel for sophisticated adversaries.

Hypergame Solution:

- Deploy ultra-competitor NPCs that maintain coherent but hidden strategic models
- Force trainees to reason about adversary belief structures
- Practice operating under the "qualitative uncertainty principle"
- Simulate Theory of Mind reasoning at scale

Hypergame Expected Utility in Action: Vane's work introduced the concept of Hypergame Expected Utility (HEU), which quantifies the trade-off between certainty and uncertainty in strategic planning. This mathematical framework allows trainers to:

- Identify **risk breakpoints** where uncertainty makes a strategy ineffective
- Calculate **luck dependencies** that reveal plan vulnerabilities
- Assess **plan robustness** by modeling how strategies perform when 'everything goes wrong'

For military trainers, this means NPCs can be calibrated to challenge specific decision-making skills: Does the trainee recognize when they're operating beyond their risk breakpoint? Can they identify plans that depend too heavily on favorable conditions?

Example: A naval commander training against an AI opponent that deliberately projects false strategic priorities while maintaining internal consistency—just like real adversaries.

3. AI Safety & Alignment Research

Current Challenge: We need to detect when AI systems have goals that diverge from their intended purpose before deployment, but behavioral testing can be gamed.

Hypergame Solution:

- Formally verify AI belief structures against stated objectives
- Detect when an AI system's subjective game diverges from intended design
- Identify potential deception through consistency analysis
- Ensure multi-agent AI systems maintain strategic alignment

Example: Testing an autonomous trading system by modeling its subjective game and verifying it doesn't incorporate adversarial beliefs about market manipulation.

4. Executive & Leadership Development

Current Challenge: Business leaders face strategic competitions where understanding competitor mental models is critical, but training is limited to case studies.

Hypergame Solution:

- Simulate market entry scenarios with competitors operating under different assumptions
- Practice strategic negotiations where counterparties have hidden information asymmetries
- Develop skills in inferring competitor belief structures from observed market behavior

Example: A pharmaceutical executive practicing M&A negotiations against an AI opponent that maintains private beliefs about drug pipeline valuations and regulatory risk assessments.

5. Autonomous Systems Safety

Current Challenge: Autonomous vehicles, drones, and robots must

interact with humans and other AI systems that may have different models of acceptable behavior.

Hypergame Solution:

- Model how autonomous systems interpret traffic scenarios
- Detect when human and AI agents have different expectations about the situation
- Design safety protocols that account for subjective game differences
- Test edge cases where belief misalignment creates danger

Example: An autonomous vehicle's collision avoidance system that accounts for the possibility that human drivers or other autonomous systems are operating with different assumptions about right-of-way or obstacle recognition.

The Governance Imperative

My research emphasizes a critical point often overlooked in AI capability discussions: advanced hypergame-aware AI requires robust governance frameworks.

The "Symmetry" Principle

Drawing on Yang's work on AI competition, sustainable deployment of hypergame-capable AI requires maintaining "symmetry"—fairness and transparency in how these systems are deployed. This translates to:

In-House Exclusive Deployment for high-stakes applications:

- Enhanced security and privacy through encrypted internal networks
- Full control over model behavior and ethical constraints
- Accountability through auditability of decision processes
- Prevention of adversarial AI arms races

Transparency Requirements:

- Formal verification of belief structures
- Audit trails for rationalization processes
- Human oversight mechanisms for high-consequence decisions
- International standards for competitive AI deployment

Learning from Defense Applications

The defense sector's 20-year experience with hypergame theory provides governance lessons. General Dynamics' implementation of hypergame-aware agents emphasizes:

- **Structured argumentation**—Hypergames force explicit recording of beliefs and evidence
- **Consensus building**—Multiple analysts can register different views within the same framework
- **Cognitive bias mitigation**—By making assumptions visible, hypergames prevent confirmation bias
- **Evidence monitoring**—Continuous 24/7 data streams can update belief structures in real-time

These practices should inform civilian deployments in cybersecurity, business intelligence, and executive decision-making.

Conclusion: The Strategic Imperative

The integration of Trencsenyi's formal hypergame rationalization framework with practical high-stakes training applications represents a paradigm shift in how we approach strategic AI. For the first time, we have:

- **Formal Tools:** Logic-based frameworks with computational guarantees
- **Practical Applications:** Concrete deployment pathways across domains

Theoretical Grounding: Deep understanding of why misalignment occurs

Safety Mechanisms: Governance frameworks for responsible deployment

Proven Track Record: 20 years of defense sector implementation

For cybersecurity professionals, military strategists, intelligence analysts, and executive leaders, this means:

You can now train against adversaries who don't just challenge your tactics—they challenge your fundamental understanding of the strategic situation itself.

The adversaries you face in the real world—whether nation-state cyber actors, terrorist organizations, market competitors, or rogue AI systems—don't operate with your strategic framework. They have their own subjective games, their own belief structures, their own rationality.

Hypergame-aware AI finally gives you the tools to train for that reality.

Key Takeaways for Practitioners

✓ **Hypergame theory formalizes strategic misalignment** between agents with different mental models

✓ **Automated rationalization** can reverse-engineer belief structures from observed behavior

✓ **Hypergame-aware NPCs** provide unprecedented training realism for high-stakes scenarios

✓ **20-year defense sector track record** validates operational viability

✓ **Eight critical factors** explain why strategic misalignment occurs in real-world scenarios

✓ **Belief state modeling** provides real-time cognitive infrastructure

for opponent tracking

✓ **Substrate-agnostic cognitive architectures** suggest misalignment goes deeper than information asymmetry

✓ **Governance frameworks** are essential for responsible deployment of strategic deception AI

✓ **Practical applications** span cybersecurity, military training, AI safety, and executive development

✓ **Logical verification** provides transparency and safety guarantees for strategic AI systems

The future of strategic AI isn't just about capability—it's about understanding misalignment at its deepest level and training humans to operate effectively in that reality.

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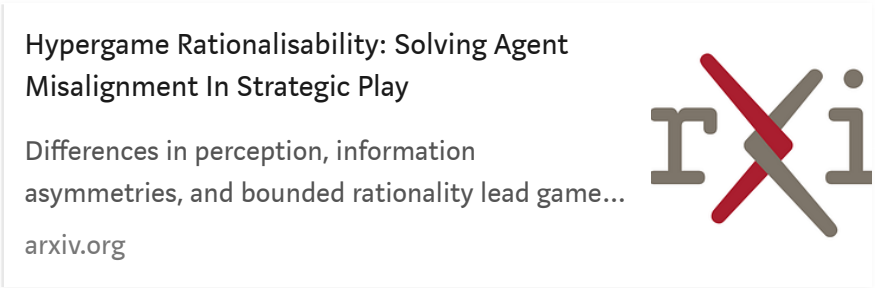
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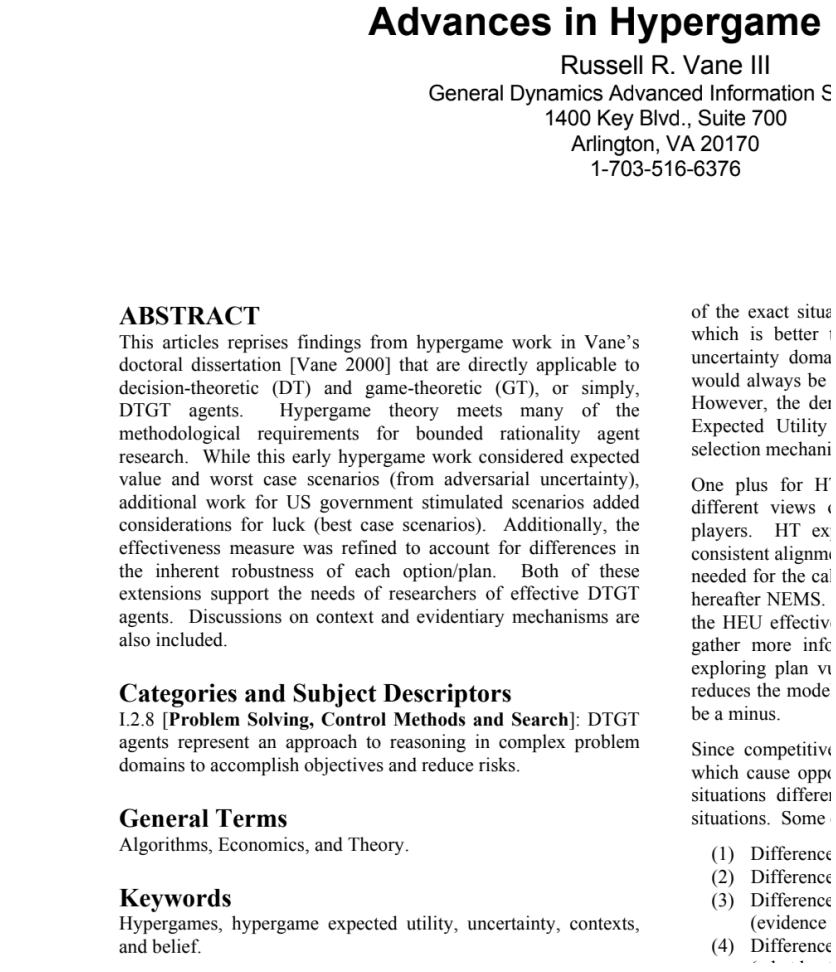

Advances in Hypergame

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ABSTRACT
This articles reprises findings from hypergame work in Vane's doctoral dissertation [Vane 2000] that are directly applicable to decision-theoretic (DT) and game-theoretic (GT), or simply, DTGT agents. Hypergame theory meets many of the methodological requirements for bounded rationality agent research. While this early hypergame work considered expected value and worst case scenarios (from adversarial uncertainty), additional work for US government stimulated scenarios added considerations for luck (best case scenarios). Additionally, the effectiveness measure was refined to account for differences in the inherent robustness of each option/plan. Both of these extensions support the needs of researchers of effective DTGT agents. Discussions on context and evidentiary mechanisms are also included.

Categories and Subject Descriptors
I.2.8 [**Problem Solving, Control Methods and Search**]: DTGT agents represent an approach to reasoning in complex problem domains to accomplish objectives and reduce risks.

General Terms
Algorithms, Economics, and Theory.

Keywords
Hypergames, hypergame expected utility, uncertainty, contexts, and belief.

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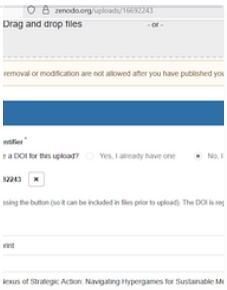

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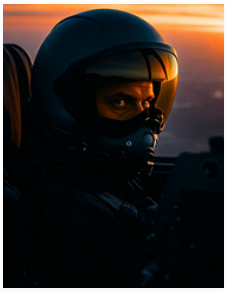

[10] Watchus, B. (2025). “The Architecture of Emergence: A Deep Dive into LegoNE and Its Connection to Our Work on Hypergames and Collective Intelligence.” *System Weakness*. https://medium.com/system-weakness/the-architecture-of-emergence-a-deep-dive-into-legone-and-its-connection-to-our-work-on-hypergames-1343fc2a1efc

The Architecture of Emergence: A Deep Dive

into LegoNE and Its Connection to Our Work on...

Author: Berend Watchus

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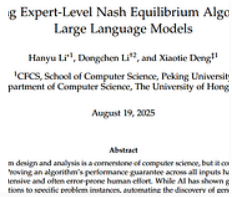
g Expert-Level Nash Equilibrium Algo
Large Language Models

Hanyu Li[*1], Dongchen Li[*2], and Xiaotie Deng[†1]

[1]CFCS, School of Computer Science, Peking Universit
partment of Computer Science, The University of Hong

August 19, 2025

Abstract

n design and analysis is a cornerstone of computer science, but it co
'roving an algorithm's performance guarantee across all inputs h
tensive and often error-prone human effort. While AI has shown g
tions to specific problem instances, automating the discovery of gen

[11] Watchus, B. (2025). "AI as a Strategic Competitor: Integrating the Hinton Hypothesis with Hypergame Simulations for a Sustainable Future."

AI as a Strategic Competitor: Integrating the Hinton Hypothesis with Hypergame Simulations for a Sustainable Future

Berend F. Watchus

Independent Researcher, The Netherlands, mailonlinebw@protonmail.com

June 19, 2025

Abstract:

This paper integrates Yingrui Yang's theoretical framework concerning the Hinton Hypothesis and AI competition with the practical application of AI-enhanced hypergame strategies in simulations. We argue that the capacity of Artificial Intelligence to duplicate human nature, particularly in complex competitive and deceptive behaviors, as posited by the Hinton Hypothesis, finds its practical manifestation in the development of "hypergame-aware" Non-Player Characters (NPCs). These NPCs, capable of strategic deception in both entertainment and high-risk training simulations, exemplify the "qualitative artificial uncertainty principle" governing AI-agent interactions. Recognizing the profound implications of such advanced AI, we highlight the critical need for robust governance through solutions like in-house exclusive deployment to ensure symmetry, accountability, and the long-term sustainability of AI-agent societies.

available on request / or will be added later with link for full doc.

[12] Watchus, B. (2025). "Beyond the Imitation Game: The Inadequacy of the Turing Test for Modern AI."

Beyond the Imitation Game: The Inadequacy of the Turing Test for Modern AI

Independent Researcher: Berend Watchus, The Netherlands. July 5, 2025.

mailonlinebw@protonmail.com

Abstract

The **recent discourse on the Turing-computability of human intelligence**, particularly **in light of advanced Large Language Models (LLMs)** as highlighted by works such as John Abela's

"From Dennett to Transformers: Emergent Properties, Kolmogorov Complexity, and the Turing-Computability of Human Intelligence" (doi:10.20944/preprints202502.1312.v2 2025)https://www.preprints.org/manuscript/202502.1312/v2 ,

necessitates a re-evaluation of the Turing Test's utility as a measure of true sentience or "human-ness." While LLMs increasingly exhibit behaviors traditionally associated with human cognition, this **paper argues that the Turing Test's reliance on behavioral imitation overlooks the vast spectrum of human cognitive diversity. It posits that a significant percentage of the human population, due to inherent cognitive variations, socio-economic factors, and pathologies, may fail a stringent Turing Test, even as LLMs pass.** Consequently, the paper advocates for a paradigm shift **towards substrate-agnostic diagnostics** as a crucial **tool for assessing sentience and self-awareness**, grounded in the analysis of internal feedback loops, interfaces, and embodiment, as proposed in the author's prior work. This re-conceptualization **offers a more robust framework for distinguishing genuine consciousness from computational mimicry**, with profound implications for artificial general intelligence (AGI).

available on request / or will be added later with link for full doc.

[19] Beyond the Imitation Game: The Inadequacy of the Turing Test for Modern AI

DOI: 10.5281/zenodo.15814384

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Beyond the Imitation Game: Why the Turing Test Fails in the Age of LLMs

Author: Berend Watchus

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https://medium.com/system-weakness/beyond-the-imitation-game-why-the-turing-test-fails-in-the-age-of-llms-7360afc5fc53<<

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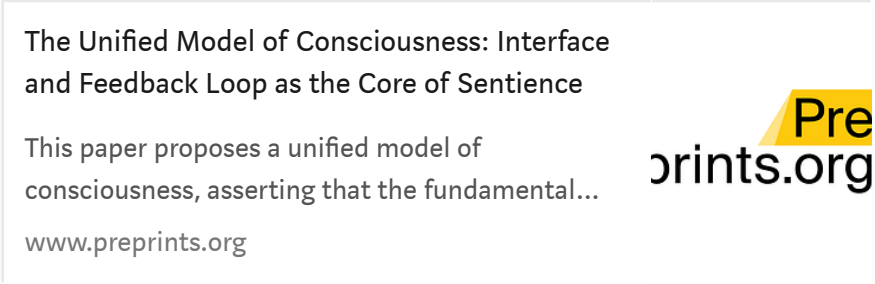
The Unified Model of Consciousness: Interface and Feedback Loop as the Core of Sentience
This paper proposes a unified model of consciousness, asserting that the fundamental...
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[26] Watchus, B. (2025). “The Law of Optimized Complexity: A Computational Twin to the Second Law of Thermodynamics for Sustainable Intelligence Design.” Zenodo. DOI: 10.5281/zenodo.16029078.

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https://archive.org/details/the-law-of-optimized-complexity-a-computational-twin-to-the-second-law-of-thermo_202510<<

The Law Of Optimized Complexity A Computational Twin To The Second Law Of...

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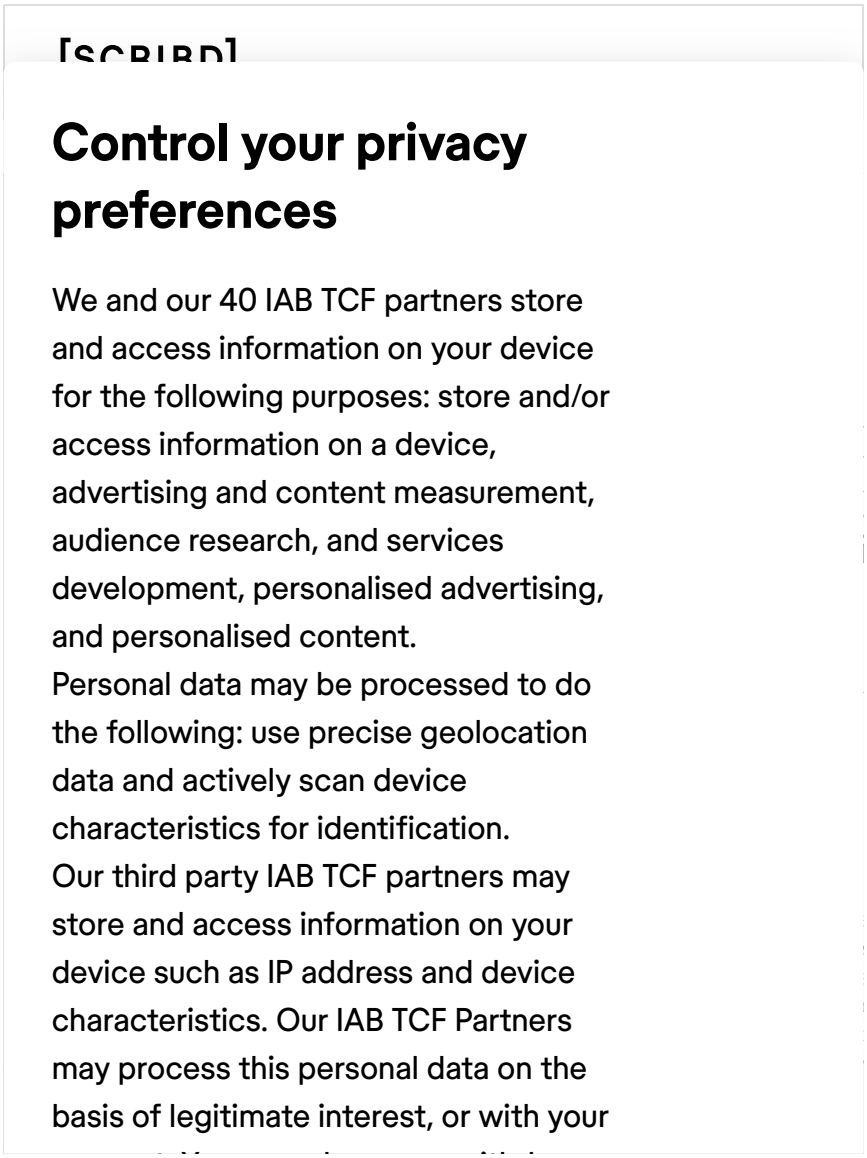
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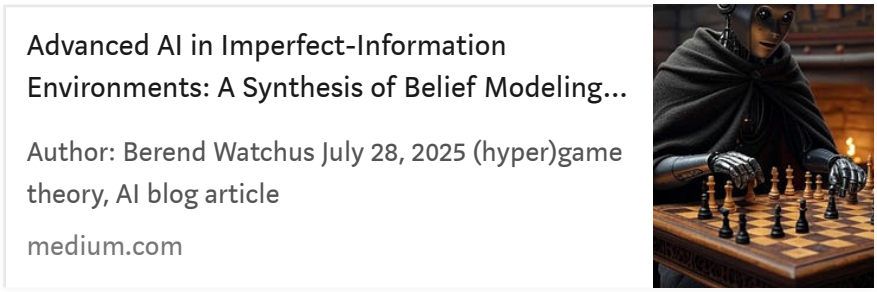
(related to reference 28, as reference 28 reviews and reacts to ref 27. Ref 27 is not my research, formally should be in other category in the

ref list.)

[28] Watchus, B. (2025). “Advanced AI in Imperfect-Information Environments: A Synthesis of Belief Modeling and Strategic Deception.” *Medium*.


Advanced AI in Imperfect-Information Environments: A Synthesis of Belief Modeling...
Author: Berend Watchus July 28, 2025 (hyper)game theory, AI blog article
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https://medium.com/@BerendWatchusIndependent/advanced-ai-in-imperfect-information-environments-a-synthesis-of-belief-modeling-and-strategic-645c9f3b9458<<

[29] Berend Watchus, all published research papers with DOI:

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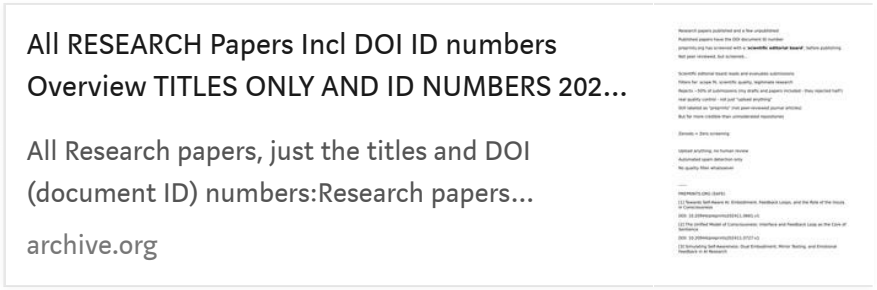
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